

# Gap Puzzles

## Puzzle solving using Prolog constraints

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**Abstract.** This project was developed during the the Logical Programming Course in the context of the year of the Integrated Masters in Informatics and Computing Engineering at The Faculty of Engineering of the University of Porto. The goal was to solve a problem using the Prolog programming language, more specifically, using Constraint Logic Programming over Finite Domains (clpfd) library, that's defined in the SICStus Prolog distribution. In the specific case of our team, the problem in question is the Gap Puzzle, which is defined in detail in **Section 2**. In order to solve the problem, Prolog constraint programming was used. This type of programming grants us the advantage of fast computation, requiring a somewhat simple and little verbosed kind of programming. Using constraints, we can simply define all the constraints of our problem (in the case of puzzles these are the rules of the puzzle), which in the case of more complex problems is not trivial. Then, after defining the solution search algorithm, the clpfd library will search for solutions.

**Keywords:** Prolog · Constraint Programing · Gap Puzzle.

## 1 Introduction

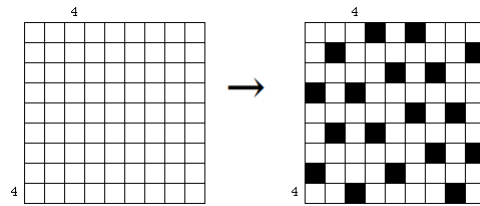
This goal of the project was to elaborate a Prolog program that, given a board order, computes all the solutions of the Gap Puzzle for that board order. Optimization is a relevant factor here. Different ways of defining constraints can lead us to the same set of solutions, but computation times can vary greatly. So, one of the concerns during the development of this project was to find solutions as fast as possible.

Bellow we shortly explain the following sections of this article.

- **Problem Description:** Detailed description of the problem we are solving and the rules of Gap Puzzles
- **Approach**
  - **Decision Variables**
  - **Constraints**
  - **Search Strategy**
- **Solution Presentation:** Description of how solutions are presented in an intelligible way.
- **Results:**
- **Conclusions:**

## 2 Problem Description

The problem in question is the solution of the Gap Puzzle. Gap is a very simple puzzle, yet a challenging one. We have grid of arbitrary size and our goal, is to shade two squares of the grid in every row and column, such that two shaded squares do not touch, even at the corners. Because there is more than one solution for every grid order, in the puzzle form of the problem, there are numbers on the side of the grid to indicate the number of non shaded squares there must exist between two shaded squares in that row or column. The figure below illustrates an example of puzzle with order 9.



**Fig. 1.** Gap Puzzle example

Although they are very interesting, these specific puzzles do not interest us. We are interested in finding arbitrary solutions for a problem with a given order.

### 3 Approach

In the specific context of this problem, the constraints used to solve the problem are a translation of the rules of the puzzle to Prolog Constraints. The following subsections explain the constraints used in detail.

#### 3.1 Decision Variables

To solve this problem we have only a decision variable, the *Points* list. This list stores the coordinates of all the cells that will be shaded, in the format  $[X1, Y1, X2, Y2 \dots]$ . We start by defining that this list will have a length of:

$$points = 2 * order \quad (1)$$

$$length = 4 * order \quad (2)$$

This happens because we can only have two shaded cells in each row. Since the number of rows is equal to the order of our grid, we have a number of points that is two times our order.

Then we define the domain of the decision variable, which is between zero and the number immediately inferior to our order, this is because we are considering a system of coordinates starting in zero.

#### 3.2 Constraints

The first constraint we make is to ensure there are only two shaded cells in each row and in each column. We do this using the **restrict\_two(+List,+Order)** predicate. This predicate restricts each number of the domain in the list to occur exactly two times in the list. We do this two times, the first for a list containing all the X coordinates of the solution, and a second time for the Y coordinate. We are able to split this using the **extractX(+List,-XList)** and **extractY(+List,-YList)** predicates. The first one takes a list with the solution format and unifies in *XList* all the X coordinates of that list. The *extractY* predicate does the same for Y.

```
restrict_two(Sol,0) :- count(0,Sol,2).

restrict_two(Sol,S) :-
    count(S,Sol,2),
    Next is S-1,
    restrict_two(Sol,Next).
```

**Fig. 2.** The restrict\_two predicate

Then we restrict the positions of the shaded cells themselves. We do this using the **restrict\_next(+List)** predicate. This predicate iterates through the list making sure that no other coordinates in the list are the coordinates of an adjacent cell to the any of the other cells and that no two cell coordinates are the same. This predicate calls the **restrict\_next\_pair(+List,+X,+Y)** predicate that applies the constraints mentioned above to all the other coordinate pairs in the list.

```
restrict_next_pair([],_,_).

restrict_next_pair([HX,HY|T],X,Y) :-
    #\((HX #= X + 1) #/\ (HY #= Y + 1)),
    #\((HX #= X + 1) #/\ (HY #= Y - 1)),
    #\((HX #= X - 1) #/\ (HY #= Y + 1)),
    #\((HX #= X - 1) #/\ (HY #= Y - 1)),
    #\((HX #= X + 1) #/\ (HY #= Y)),
    #\((HX #= X - 1) #/\ (HY #= Y)),
    #\((HX #= X) #/\ (HY #= Y - 1)),
    #\((HX #= X) #/\ (HY #= Y + 1)),
    #\((HX #= X) #/\ (HY #= Y)),
    restrict_next_pair(T,X,Y).
```

**Fig. 3.** The restrict\_next\_pair predicate

### 3.3 Search Strategy

For labeling we made an ordering function in **crescent\_order(+List)** predicate. This predicate orders the coordinates of the points down-top and right-left, improving the labeling times for smaller boards.

```
crescent_order([_,_]).

crescent_order([X1,Y1,X2,Y2|T]) :-
    X1 #>= X2,
    #\((X1 #= X2) #<=> (Y1 #> Y2)),
    crescent_order([X2,Y2|T]).

crescent_order([H1,H2|T]) :-
    H1 #>= H2,
    crescent_order([H2|T]).
```

**Fig. 4.** Point ordering

For bigger boards, **crescent\_order(+List)** predicate is an expensive task, as it runs through the whole list of points to apply the restrictions. For this reason, if the board is of order 15 or bigger, this predicate is not used.

We also added a random function that was given in the classes. This function allows **labeling** predicate to select the value in a random way.

```
labeling([value(selRandom)],Points),
statistics(walltime, [End,_]),
display_gap(Points,S),

Time is End - Start,
format('finished ~3d seconds.~n', [Time]).

selRandom(Var, Rest, BB0, BB1):- % seleciona valor de forma aleatória
    fd_set(Var, Set), fdset_to_list(Set, List),
    random_member(Value, List), % da library(random)
    ( first_bound(BB0, BB1), Var #= Value ;
      later_bound(BB0, BB1), Var #\= Value ).
```

Fig. 5. Random selection predicate

## 4 Solution Presentation

For solution presentation we elaborated **display\_gap(+Sol,+Len)**. This predicate receives the *Points* list, with the format previously explained, and *Order*, which is the order of the board the solution given applies to. This predicate then uses the three predicates explained below.

Firstly it uses **empty\_board(-B,+Len)** predicate which creates a list of lists representing an empty grid, with order *Len*, this new grid is then unified with *B*. This predicate works by appending new atoms, representing an empty cell ( ' ', a single space character) to an empty starting list. Once this list has the desired length, this list is then replicated several times to form a new empty grid.

Then, we use the **process\_board(+Board,+Sol,-NewBoard)** to substitute all the empty cells in the positions given by solution list, with the ' X ' atom. This represents a shaded cell. In order to make the substitution we defined the **replaceN(+List,+N,+NewElem,-Res)** predicate. This predicate substitutes the element of order *N* in *List* to the element *NewElem*, then unifies the resulting list with *Res*.

Finally, the predicate **display\_board(+Board)** was defined to display the board *Board*. This predicate writes the separators in the screen and then displays each row of the grid, using the right format. Below, we can see a displayed board.

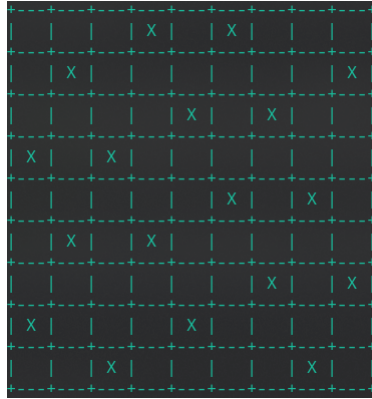


Fig. 6. Solution Display Example

## 5 Results

Our program finds one solution for every order given. Given this, we tested it with all possible numbers from 1 to 30. By doing this we discovered that the puzzle only has solutions with orders 9 and above. Here is one example of one board with order 11.

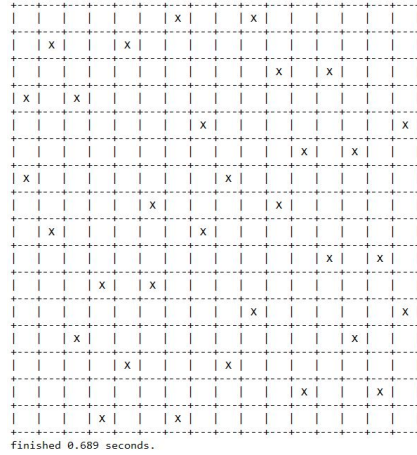
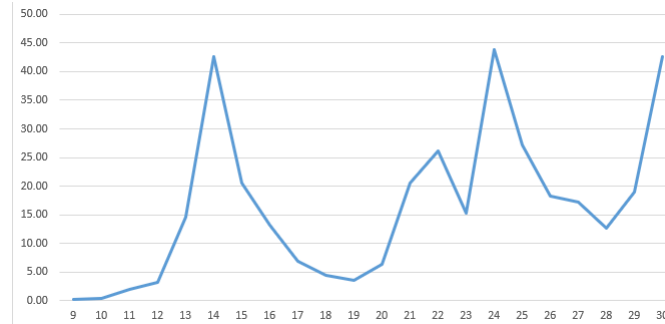


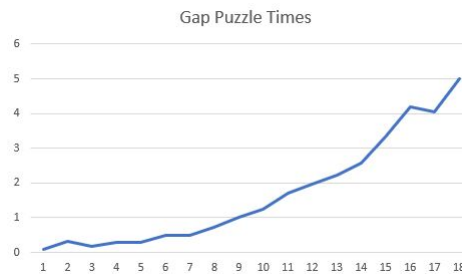
Fig. 7. Solution Example

After making multiple tries per order we ended up with this graph. It shows the amount of time (in seconds) taken by the program for each order. From the graph you can see a pattern where the program takes exponentially more time to solve the puzzle. But then it reaches a point where there is a drop in that time. This pattern repeats itself after its first instance and probably forever, as you increase order.



**Fig. 8.** Times graph

This irregular behaviour told us the solution was imperfect somewhere. With that knowledge, we felt incentivized to search new ways of labeling variables. We then found we could add a random function to the labeling and once we implemented it, it showed incredible results. After more testing, we reached the conclusion that the best way to search for solutions in smaller boards was to use **crescent\_order(+List)**, but to bigger boards it was a waste of time. With this in mind, we added a condition where if the board was of order 14 or below, it would use **crescent\_order(+List)**, but bigger boards it would not.



**Fig. 9.** Times graph after random labeling

## 6 Conclusions and Future Work

We think we have achieved what was proposed to us at the start of this project. With this work we were able to learn how to think in a different way. Since we use other paradigms of programming more frequently, when using prolog with restrictions we could not think of a solution to the problem as quickly as normal.

As previously said, the gap puzzle is a challenging one, but with the use of restrictions, we can find solutions in little time. One significant thing we learned was that a change that might seem simple, as adding a random function, might increase efficiency by many times.

From the results we can conclude that our program is quite efficient and that is one of its advantages. But this is at the cost of certainty, since you only get random solutions, maybe not the one you are looking for.

## References

1. Erich Friedman's Puzzle Palace - Gap Puzzles, Copyright Erich Friedman, 2010. <https://www2.stetson.edu/~efriedma/puzzle/gap/>. Last accessed 4 Jan 2020